

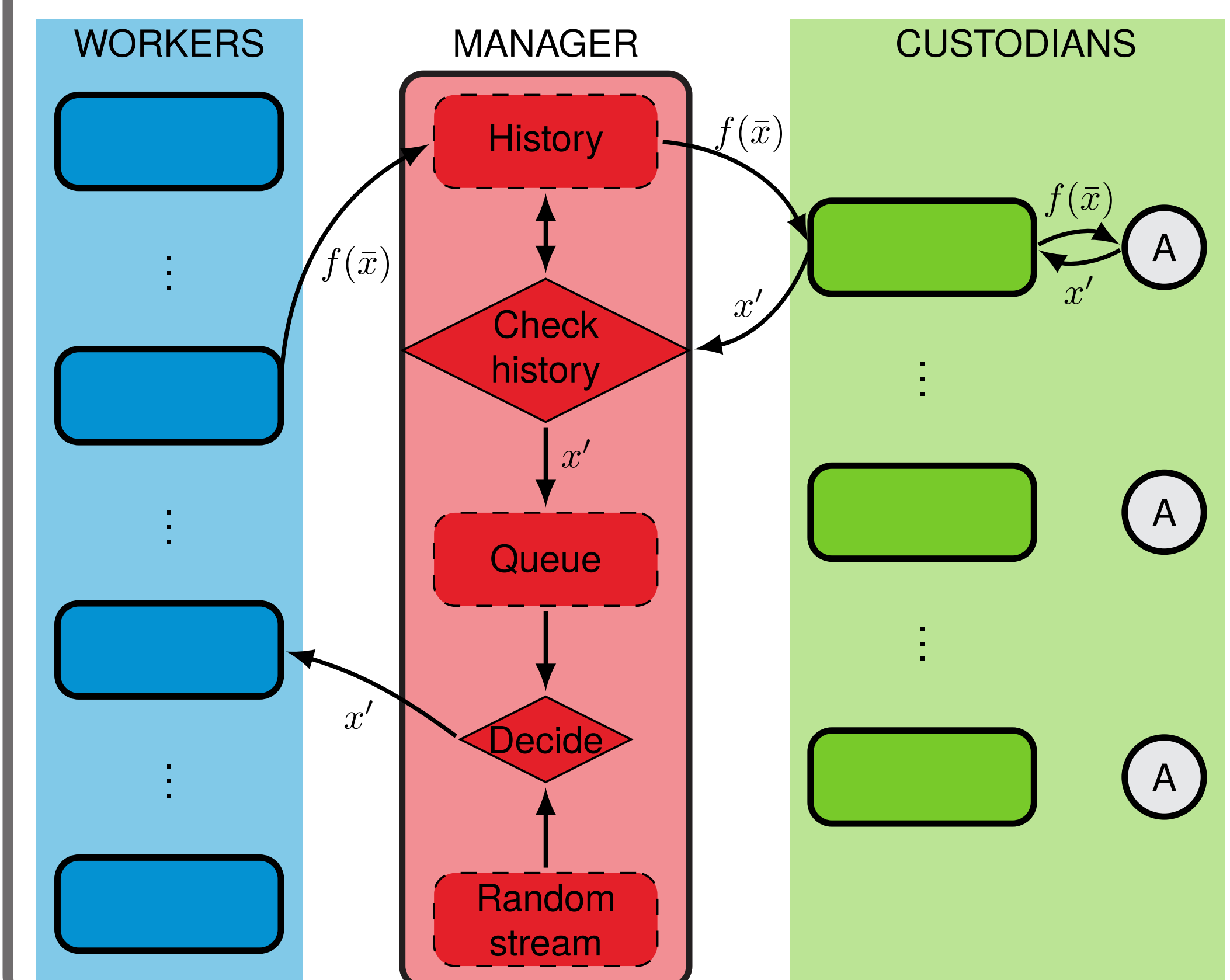
Problem statement

We want to find multiple, high-quality local minima of the nonlinear optimization problem

$$\begin{aligned} &\text{minimize } f(x) \\ &\quad x \in \mathbb{R}^n \\ &\text{subject to: } x \in \mathcal{D} \end{aligned}$$

when \mathcal{D} is compact, concurrent evaluations of f are possible, and relatively little is known about f .

Asynchronous workflow



APOSMM

input: Local optimization method, random stream \mathcal{R}_S , tolerance ν .

for $w = \{1, \dots, c\}$ **do**

 Give w a point from \mathcal{R}_S at which to evaluate f .

for $k = 0, 1, \dots$ **do**

 Receive from worker w that has evaluated its point \tilde{x} .

if $\tilde{x} \in A_k$ **then**

if \tilde{x} 's run is complete **then**

 Add minimizer to X_k^* ; remove points from run from A_k .

else

 Query local optimization method and add subsequent point from \tilde{x} 's run to Q_L .

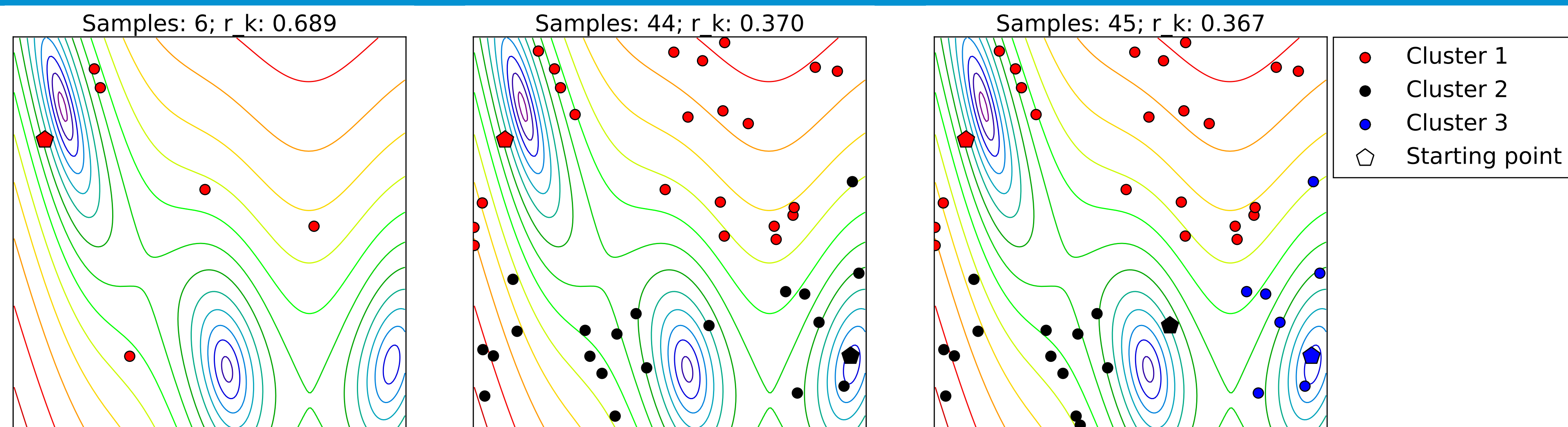
 Update H_k ; Update r_k using (1).

 Start local optimization method at points in H_k satisfying certain conditions; add the subsequent point(s) to Q_L .

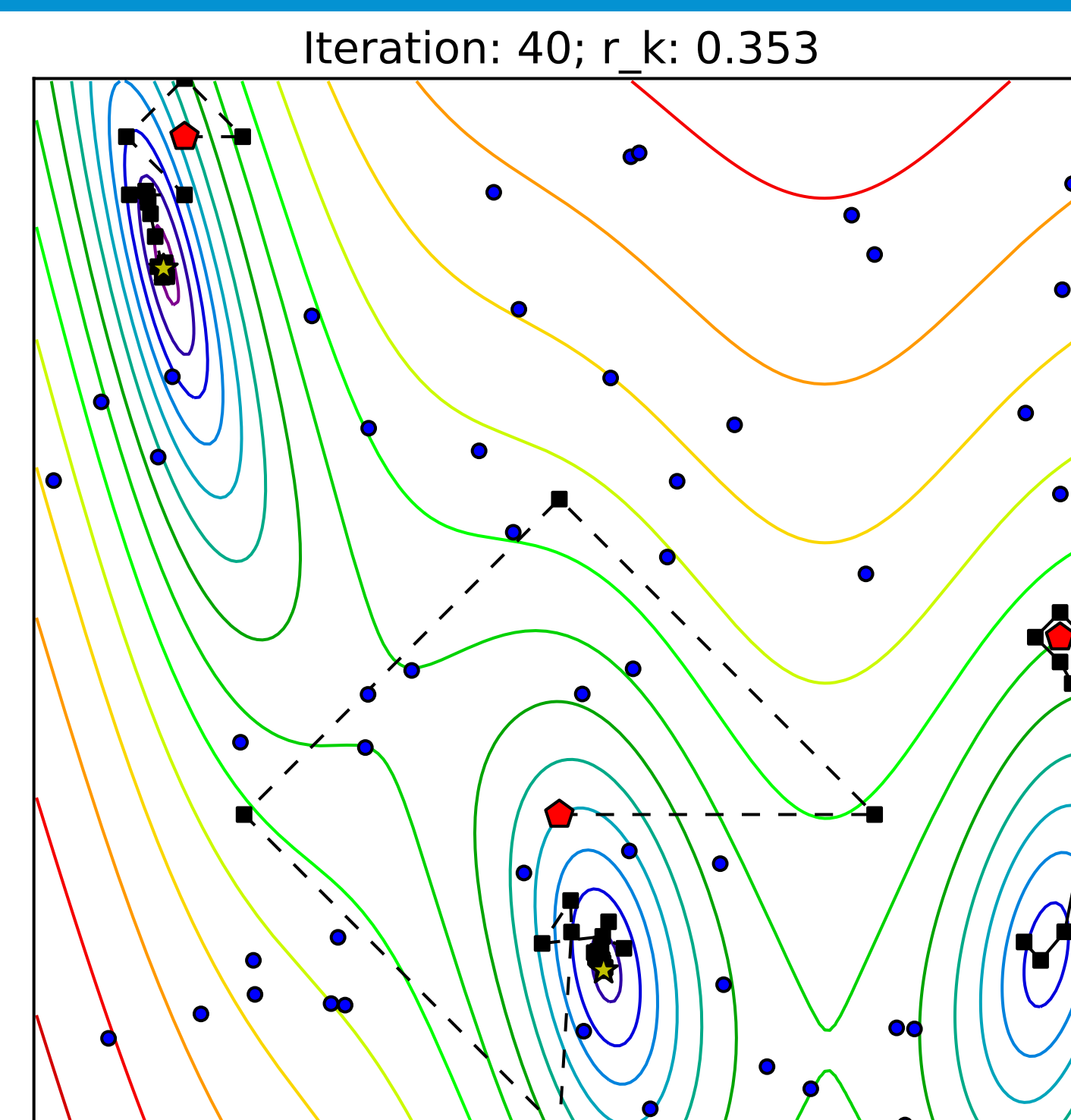
 Kill runs with candidate minima within 2ν of each other, keeping the best run.

 Give w a point x' at which to evaluate f , either from Q_L or \mathcal{R}_S .

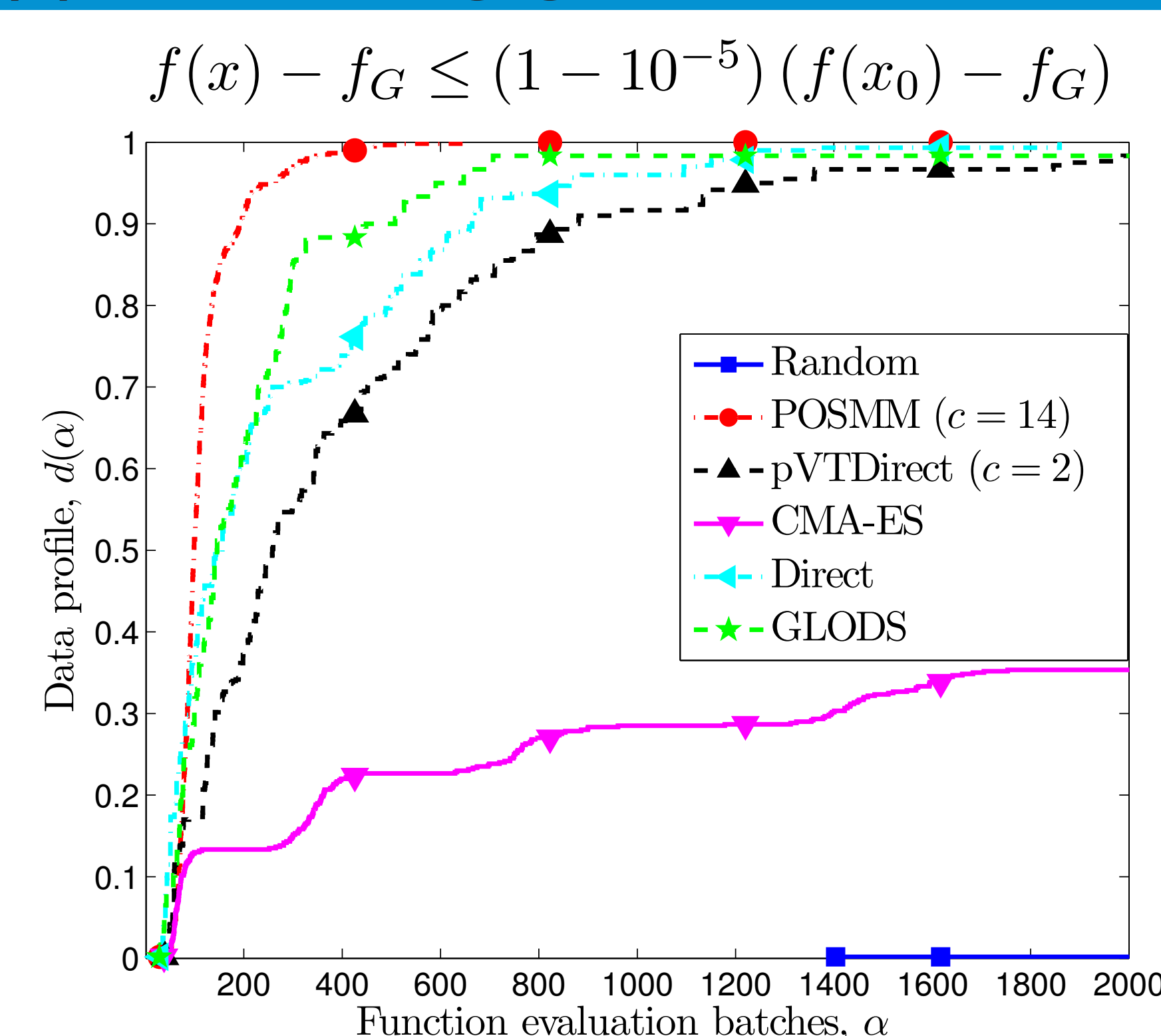
Conditions



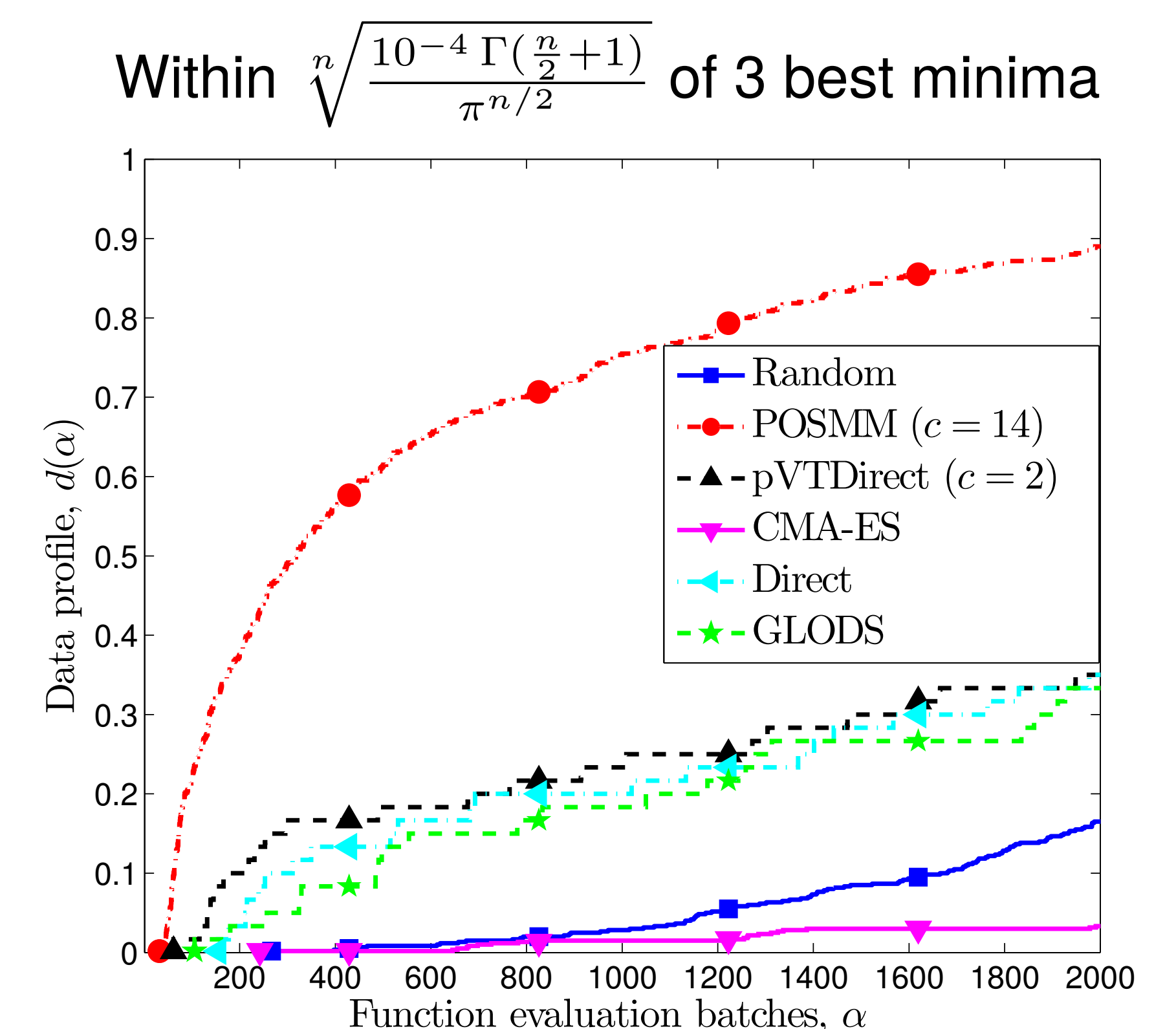
Snapshot



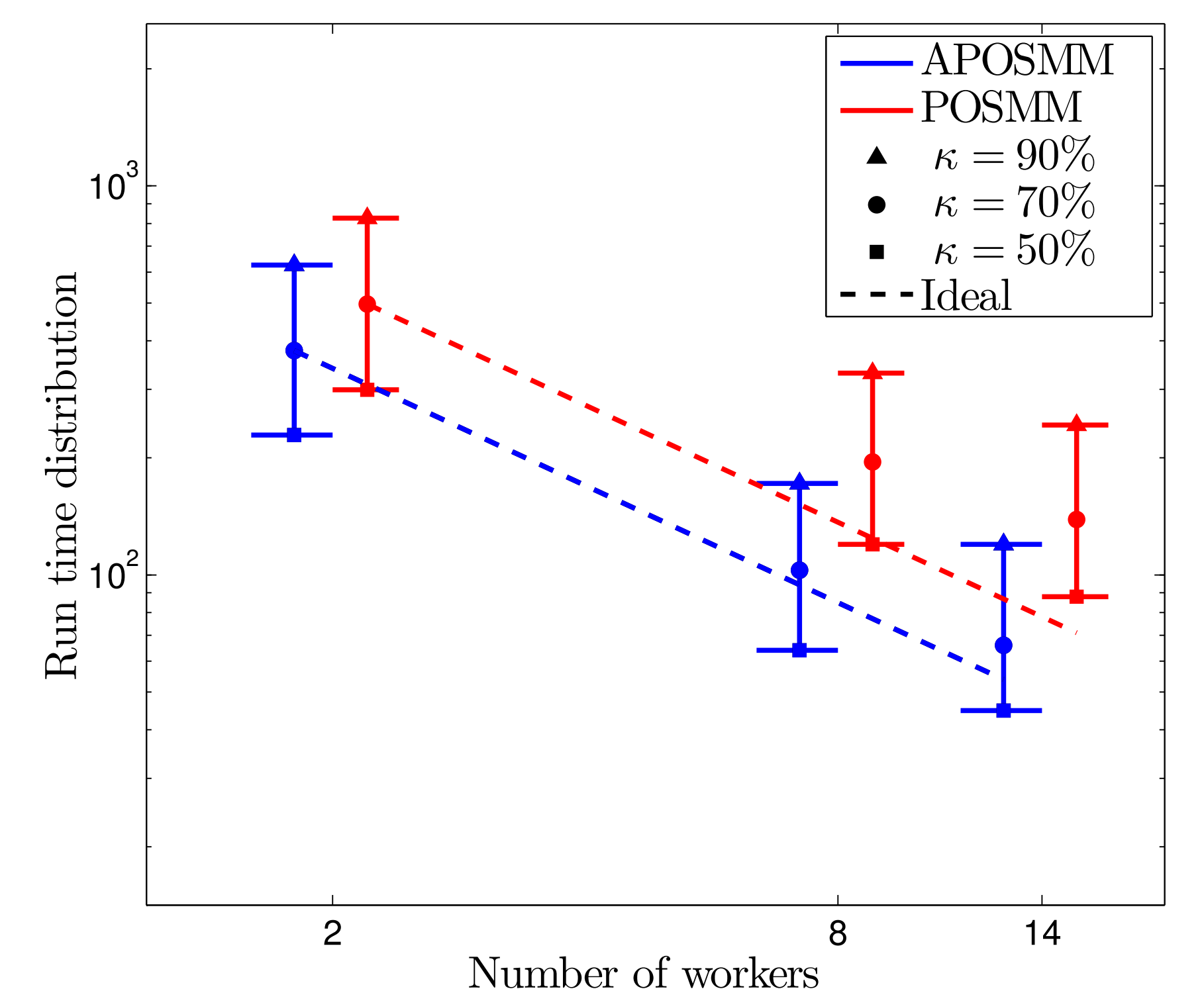
Approximating global minima



Finding multiple minima



Performance scalability



Highlights

- Multistart algorithm that considers all previously evaluated points when deciding where to start or continue a local optimization run.
- Applicable to general optimization; its judicious use of function evaluations is especially suited for expensive derivative-free objectives.
- Time to solution scales well even when the time to evaluate the objective is highly variable.
- The algorithm has strong theoretical properties and performs well in practice.
- Depends on the critical distance

$$r_k = \frac{1}{\sqrt{\pi}} \sqrt[n]{\Gamma\left(1 + \frac{n}{2}\right) \text{vol}(\mathcal{D}) \frac{5 \log |\mathcal{S}_k|}{|\mathcal{S}_k|}}. \quad (1)$$

Theorem

- If $f \in C^2$,
- there is a distance $\epsilon > 0$ between local minima,
- the local optimization method is strictly descent,
- r_k is defined by (1), then

APOSMM almost surely starts a finite number of local optimization runs and every local minimum is found or has a single local optimization run asymptotically converging to it.

J. Larson and S. M. Wild. Asynchronously parallel optimization solver for finding multiple minima. ANL/MCS-P5575-0316, 2016

J. Larson and S. M. Wild. A batch, derivative-free algorithm for finding multiple local minima. Optimization and Engineering 17(1), 2016